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Local versus broad scale environmental drivers of continental beta diversity patterns in subterranean spider communities across Europe

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Running head: Continental patterns of cave spiders

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50 ABSTRACT

51

52 Macroecologists seek to identify drivers of community turnover (β -diversity) through broad spatial
53 scales. Yet, the influence of local habitat features in driving broad-scale β -diversity patterns
54 remains largely untested, due to the objective challenges of associating local-scale variables to
55 continental-framed datasets. We examined the relative contribution of local- versus broad-scale
56 drivers of continental β -diversity patterns, using a uniquely suited dataset of cave-dwelling spider
57 communities across Europe (35–70° latitude). Generalized dissimilarity modeling showed that
58 geographical distance, mean annual temperature, and size of the karst area in which caves
59 occurred drove most of β -diversity, with differential contributions of each factor according to the
60 level of subterranean specialization. Highly specialized communities were mostly influenced by
61 geographical distance, while less specialized communities were mostly driven by mean annual
62 temperature. Conversely, local-scale habitat features turned out to be meaningless predictors of
63 community change, which emphasizes the idea of caves as the human accessible fraction of the
64 extended network of fissures that more properly represents the elective habitat of the subterranean
65 fauna. To the extent that the effect of local features turned to be inconspicuous, caves emerge as
66 experimental model systems in which to study broad biological patterns without the confounding
67 effect of local habitat features.

68

69 **Keywords:** Araneae, cave, Europe, generalized dissimilarity model, latitudinal gradient, subterranean
70 biodiversity

71

72

73 BACKGROUND

74

75 Understanding why biological communities differ from one another is among the most basal
76 research questions in ecology, yet answering this question represents a significant intellectual
77 challenge (1,2). For over a century, species richness (α -diversity) has been the most commonly
78 used metric to quantify and explore biological diversity through the environmental space (3). Yet it
79 is increasingly acknowledged that the extent of change in community composition along gradients
80 (β -diversity) is a prominent and complementary feature to consider as well, possibly even more
81 meaningful than α -diversity when dealing with macroecological patterns (4). Substantial turnovers
82 in the composition of communities along broad-scale ecological gradients have been observed in
83 virtually all taxa (5). Community changes following latitudinal clines or elevational extents (that is,
84 essentially thermal seasonality gradients) (6–9), gradients of productivity (10), urbanization (11,12),
85 or salinity (13,14), are all examples in which one or a few broad-scale environmental gradients well
86 explained turnover in biological communities. Yet an equally important, but potentially
87 inconspicuous and overlooked, facet of β -diversity analyses pertains the contribution of local-scale
88 ecological factors in explaining patterns of biological diversity at a broader spatial scale (15).
89 Accounting for local features like microhabitat characteristics (16,17) and local land use (18) might
90 provide complementary information to understand the ecological processes involved in filtering
91 larger species pools to the subset of resident species that occurs within a given community.
92 However, the objective challenge of associating local environmental and habitat features to
93 continental and global biodiversity datasets has largely prevented macroecologists to incorporate
94 local-scale features in their modeling exercises.

95 Here we use a uniquely suited continental dataset of cave-dwelling spider communities to
96 examine the relative contribution of local- versus broad-scale predictors—i.e., the local
97 geomorphological features of caves versus broad-scale environmental predictors commonly used
98 in macroecological analyses—in driving community turnover through space. Whilst subterranean
99 habitats have largely been omitted in exploring β -diversity patterns across continental scales
100 (19,20), cave communities provide the discrete boundaries and simplified biological assemblages
101 which are often required for similar analyses (21). The fact that cave systems are extensively
102 replicated across the Earth (22) offers the unique opportunity of studying semi-closed habitats
103 characterized by relatively homogenous and recurrent structural characteristics distributed along
104 broad-scale gradients of varying climatic conditions. A number of studies demonstrated the
105 importance of local features in determining subterranean species richness at the level of a single
106 cave or a few karst systems (23–27), suggesting that the signature of the environmental filtering
107 posed by local habitat features could potentially be detected also at broader scales (28).

108 Gauging the relative contribution of local- and broad-scale drivers of macro-diversity
109 patterns is challenging not only due to the general lack of suitable datasets for performing similar
110 tasks, but also because the organisms interacting within a typical community—even in a cave—are
111 often spectacularly polyphyletic and functionally diverse (22). To minimize noise, it is thus
112 convenient to focus on specific model organisms deemed to be good representatives of the
113 response of the biological communities due to their clear and specific ecological role. Among other
114 subterranean components, spiders (Arachnida: Araneae) are widespread and distinctive for their
115 key role as predators in the subterranean trophic webs (29). When accounting for the nearly 500
116 spider species inhabiting subterranean habitats in Europe, there are species with different levels of
117 specialization and affinity to the subterranean environment, from obligate cave-dwellers
118 (troglobionts) to species with only partial affinity to caves (troglophiles) (30). This great diversity
119 offers a wide analytical spectrum, insofar as differences in subterranean specialization and
120 dispersal propensity might lead to diverse distribution patterns.

121 We assembled a dataset of 475 subterranean spider communities across Europe thanks to
122 the effort of an international network of araneologists, biospeleologists and cavers. This dataset is
123 unique in that it covers a large geographical extent on the one hand, and contains high-resolution
124 local data on geomorphological and habitat features on the other. Analyzing it by using generalized
125 dissimilarity models (31), a novel modeling technique that accommodates for nonlinearity and non-
126 stationarity in matrix regressions, we explored the following questions:

- 127
- 128 i) What is the contribution of local factors versus broad-scale environmental factors in determining
129 community turnover among cave spider communities across Europe?
 - 130
 - 131 ii) Are the observed patterns influenced by the level of subterranean specialization of the different
132 species (that is, troglobiont versus troglophile spider assemblages)?

133 METHODS

134 Dataset assemblage

135 We compiled what we believe to be the first continental-scale geo-referenced dataset of
136 subterranean spider communities across Europe (32). The dataset comprises data from 475 caves
137 from 27 European countries, and covers a latitudinal range from 35° to 70° (Figure 1). In
138 constructing the dataset, we deliberately choose caves for which we deemed the spider fauna to
139 be exhaustively known and for which the morphological and environmental features were available,
140 thus minimizing the number of missing data ('NA') in the dataset. Although we acknowledge that
141 different sampling bias exists when it comes to estimate the diversity of species within
142 subterranean habitats (33), by selecting only well-studied caves we assumed the sampling bias to
143 be homogeneous within the caves included in the dataset.

144 To capture the diversity of subterranean habitats across Europe, the selection of the sites
145 was driven by the necessity to maximize the ranges of environmental gradients therein. First, in
146 order to account for the wide variety of habitats inhabited by subterranean spiders (29), we
147 considered as individual sites different types of caves: limestone, volcanic, talus, and salt caves,
148 but also artificial sites such as mines, blockhouses, and cellars. The general term 'cave' is used
149 hereafter. Furthermore, we selected cave openings in different types of habitats and substrates, at
150 different elevations (0 – 2,000 m a.s.l.), covering a wide range of linear planimetric development
151 (3.5 – 70,000 m), prevalent drops (from –877 to +815 m), and main entrance sizes (0.1 – 45,000
152 m²). Spatially, we selected caves so to cover the study area as homogeneously as possible. Yet,
153 the need of choosing only well-studied caves and the often clumped distribution of caves within
154 karst areas (34,35) prevented us to obtain a fully homogeneous distribution of sites (Figure 1).

155

156 Spider composition and environmental gradients

157 We associated high-resolution data to each cave, namely spider community composition along
158 with information on local geomorphological and environmental features. Spider community
159 composition was represented as incidence data—presence/absence of both described species and
160 species under description. To evaluate if drivers of β -diversity varied depending on the
161 subterranean specialization of different species, we classified each species as either 'troglophile' or
162 'troglobiont'. In subterranean biology, the term 'troglophile' is used to refer to species that are able
163 to complete their life cycles both in the subterranean and the surface environments, often forming
164 populations in both habitats. Conversely, the term 'troglobiont' refers to species that are obligate
165 subterranean dwellers (36). We use the partitioning of European spiders into these two classes
166 found in the checklist of subterranean spiders (30); species not included in the checklist were
167 classified using the same criteria. When lacking information on the distribution, habitat preference
168 and autoecology (36), the classification of a species into these two categories was based on

169 morphological traits associated with the subterranean life—depigmentation, leg elongation, and
170 eye regression. Morphological traits were derived from species descriptions, taking advantage of
171 the fact that taxonomic literature on spiders is fully digitalized and freely available online (37).
172 Accidental species, i.e., surface species not showing any morphological adaptation or association
173 with the subterranean habitats, were not included in the dataset.

174 We used as local-scale predictors all the geomorphological features of the different caves,
175 namely the elevation, number of entrances, the main entrance size (a numerical estimation of the
176 dimension of the main entrance in square meters), cave development (total planimetric
177 development of the cave in meters), prevalent drop (total positive minus total negative drop in
178 meters), as well as additional categorical features (type of cave, geological substrate, presence of
179 a subterranean river, entrance habitat, touristic use). We extracted broad-scale predictors for each
180 locality from environmental rasters at a resolution of 2.5 minutes. Climatic data were derived from
181 WORLDCLIM2 (38): mean annual temperature, annual range of temperature, cumulative
182 precipitations, and solar radiation. The reliability of these surface variables as surrogates for
183 subterranean conditions has been extensively discussed elsewhere (20,39,40). To consider the
184 possible effects related to the biogeographical history, we further included the distance from the
185 last glacial maximum (LGM) glacier as an additional broad-scale predictor. We constructed this
186 raster by buffering the shapefile of LGM glaciers with distance rings of 5 km (41). Furthermore, a
187 shapefile of carbonate extent for the study area was obtained from the World Map of Carbonate
188 Rock Outcrops (version 3.0). We rasterized the shapefile and calculated the area of each karst
189 patch (Karst area; see Figure 1). We assigned to each raster pixel the area value of the
190 corresponding karst patch (value of 0 for non-karst pixel). [A full description of all variables is given](#)
191 [in Appendix S1.](#)

192

193 **Statistical analyses**

194 We used a Wilcoxon rank sum test with continuity correction to compare median values of richness
195 of troglophile and troglobiont spider communities. To compare β -diversity, we made pairwise
196 comparisons of the 475 cave communities and computed a [Sørensen](#) dissimilarity index. We used
197 generalized dissimilarity models (GDMs) to compare patterns of β -diversity between communities
198 of troglophile and troglobiont spiders and evaluate the relative contribution of local- versus broad-
199 scale environmental gradients in explaining these patterns. GDM represents a nonlinear extension
200 of a traditional distance approach of matrix regression. It permits to analyze patterns in the
201 compositional dissimilarity among sites and to quantify how much sites differ in their environmental
202 conditions (environmental distance) and how isolated they are from one another (geographical
203 distance). In contrast to standard linear matrix regressions, a GDM accommodates for the variation
204 in the rate of compositional turnover (non-stationarity) at different positions along a given gradient,

205 and nonlinear relations between compositional dissimilarity and both environmental and
206 geographical distances between sites (31).

207 We performed matrix regressions in R (version 3.5.1) with the functions available in the
208 'gdm' package (42). We used as input data site-by-environment and site-by-species matrices for
209 troglophile and troglobiont spiders (doi:10.5061/dryad.qz612jm8z). Prior to model fitting, we
210 performed data exploration in order to detect outlying observations in the dataset and to evaluate
211 collinearity among predictors. We graphically explored the presence of outliers using Cleveland's
212 dot plots. We calculated pairwise Pearson's correlations to detect collinearity among predictors
213 using a standard $|r| > 0.70$ threshold to cull variables (43). We also used boxplots to graphically
214 assess collinearity between continuous and categorical variables.

215 We fitted individual GDMs for troglophile and troglobiont β -diversity matrices with default
216 parameters of three I-splines per predictor and knot values of 0 (minimum), 50 (median), and 100
217 quantiles (maximum). Models were weighted by species richness—for troglophiles, we also filtered
218 caves with less than four species to avoid sampling artifacts. We quantified variable importance
219 and significance using Monte Carlo matrix permutation (31,44). We retained in the final GDMs only
220 predictors that explained model variance. We plotted the I-splines of significant predictors to
221 assess how magnitudes and rates of species turnover varied along and between gradients and
222 how these patterns differed between troglophiles and troglobionts. We estimated confidence
223 intervals around the fitted I-splines using bootstrapping. Finally, to visualize multi-dimensional
224 biological patterns of β -diversity in the environmental space, we used a principal component
225 analysis (PCA) to reduce dimensionality among predictors and assigned the first three PCA
226 components to a RGB color palette.

227 RESULTS

228

229 European spider assemblages by numbers

230 The site-by-species dataset consisted of 475 caves and 331 unique species, of which 132 were
231 troglobionts and 199 were troglophiles. This diversity accounts for nearly 70% of the subterranean
232 spider species reported to occur in Europe, i.e., 486 species (30). The overall number of species
233 per cave community ranged from 0 to 15 (mean±sd = 4.44±2.25). The number of troglophile
234 species ranged from 0 to 11 (mean±sd = 3.84±2.10), whereas troglobiont species were numerically
235 lower (range = 0–5; mean±sd = 0.60±0.99) (Figure 2A). The median number of species per cave
236 was significantly higher in troglophile rather than troglobiont spider communities (Wilcoxon rank-
237 sum test, $W = 14927$, $p < 0.01$).

238 β -diversity between caves was generally higher for troglobionts than troglophiles (Figure
239 2B). The β -diversity values for troglophiles were mostly concentrated between 0.6 and 0.8. Lower
240 values were mostly due to the presence of a few widespread troglophile species, namely *Metellina*
241 *merianae* (Scopoli) (Tetragnathidae), present in 50% of the considered caves ($n = 238$), *Meta*
242 *menardi* (Latreille) (Tetragnathidae) (30%; $n = 147$), *Tegenaria silvestris* L. Koch (Agelenidae)
243 (19%; $n = 92$), and *Porrhomma convexum* (Westring) (Linyphiidae) (18%; $n = 86$). Values of β -
244 diversity for troglobionts approached 1 in most cases (Figure 2B). On average, each troglophile
245 species appeared in nine caves (mean±s.d.= 9.16±27.08; range= 1–238). Conversely, troglobiont
246 species rarely occurred in more than two caves (mean±s.d. = 2.15±2.70; range = 1–21).

247

248 Drivers of β -diversity

249 As a result of data exploration, we log-transformed the cave development and main entrance size
250 variables to homogenize their distribution and account for a few outliers. We found mean annual
251 temperature to be collinear with elevation ($|r| = 0.7$) and solar radiation ($|r| = 0.8$), hence we culled
252 the latter predictors. Mean annual temperature was also correlated with entrance habitat—caves in
253 forested areas generally displayed lower temperatures than caves opening in shrubs, grass and
254 rocky habitats—and hence we also excluded the latter categorical predictor. As a large proportion
255 of caves in the dataset were formed in limestone rocks ($n = 411$), we found the levels of the
256 categorical variable type of cave and geological substrate to be unbalanced and the variables to be
257 correlated with karst area. Thus, we dropped these predictors. We also excluded the presence of a
258 subterranean river and the touristic use of cave variables due to a significant unbalance between
259 the distribution of the observations at the two levels of these factors. Yet, a preliminary exploration
260 with a chi-squared test revealed no difference in spider richness relative to these factors [presence
261 of subterranean river: $\chi^2(13, N = 422) = 13.90$, $p = 0.91$; touristic use of cave: $\chi^2(13, N = 455) =$
262 12.91 , $p = 0.50$]. The list of local and regional predictors used in the GDMs and their significance is

presented in Table 1, whereas the full list of predictors is reported in Appendix S1.

In both troglaphiles and troglobionts, patterns of species turnover varied by environmental gradients and geographical distance. Yet, the percentage of variance explained by the models differed considerably between the two groups (18% for troglaphiles and 43% for troglobionts). Spatially, community turnover was greater in southern than northern Europe for both troglaphiles (Figure 3A) and troglobionts (Figure 4A). Troglobiont spider assemblages in the Dinaric karst (Balkan Peninsula) and Turkey emerged as the most unique. The assemblages from the Alps and the Iberian Peninsula were in general more similar. Communities at northern latitudes were the most homogeneous. In the case of troglobionts, northern communities virtually consisted of a single species, *Porrhomma rosenhaueri* (L. Koch) (Linyphiidae). For troglaphiles, mean annual temperature (Figure 3C) was the most important gradient for determining community turnover, followed by karst area (Figure 3D) and geographical distance (Figure 3E). The rate and magnitude of turnover along the gradients was exponential for mean annual temperature and linear for karst area. Turnover also increased nonlinearly with geographical distance, without reaching an asymptote. The contribution of additional drivers, both local and broad, was negligible. Predictors identified as significant by the GDM for troglobionts were, in order of importance, geographical distance (Figure 4C), mean annual temperature (Figure 4E) and karst area (Figure 4D). The model excluded other local and regional predictors (Table 1). The rates and magnitude of turnover along the geographical distance gradient were nonlinearly asymptotic, with rates of turnover steeply increasing up to 8° when they reach a plateau of full community dissimilarity (Figure 4E). The rates and magnitude of turnover of mean annual temperature and karst area were exponential. In the latter case, the contribution of additional drivers was also negligible.

285 DISCUSSION

286 Spider species richness (α -diversity) in European caves was generally low, with the majority of
287 species being distributed in one or a very few caves (Figure 2A). Number of troglobionts per cave
288 was consistently lower than number of troglophiles, an expected pattern that both reflects the
289 limited dispersal ability of troglobiont spiders (29) and the reduced availability of trophic resources
290 in the deep and inner areas of caves where they usually reside (21,22). Yet, it must be kept in mind
291 that α -diversity values are expected to be higher—and number of caves per several species to be
292 lower—than those reported here, as recent molecular studies revealed that cryptic diversity in
293 subterranean lineages is often high (45–47). Interestingly, it was demonstrated that this is not a
294 critical shortcoming in subterranean macroecological studies, since cryptic species diversity should
295 be homogeneously distributed along environmental gradients (47).

296 We observed that spider communities progressively became more homogenous from the
297 south to the north for both troglophiles (Figure 3A) and troglobionts (Figure 4A), a typical pattern in
298 the Northern Hemisphere (9,48). The same environmental gradients explained β -diversity
299 variations for both categories of taxa: geographical distance, temperature, and availability of karst.
300 Nevertheless, the relative importance of these three drivers differed substantially depending on the
301 level of subterranean specialization. Also, model fit was significantly better in the case of
302 troglobionts, whereas ~80% of the model deviance for troglophiles remained unexplained. This
303 latter group comprises a great variety of species, highly diverse in term of Linnean distance, but
304 also in morphological and life-history traits (29,30,36). An ongoing collection of subterranean spider
305 traits will possibly allow to subdivide troglophiles into more coherent functional subgroups, hence
306 increasing the explanatory power of the models.

307 Rather than environmental distance, geographical distance emerged as the most important
308 factor explaining β -diversity patterns in troglobiont spider communities. Across our study area, two
309 randomly sampled caves are predicted to be fully dissimilar in terms of community composition if
310 they are at a distance $> 8^\circ$ (Figure 4C). This result is consistent with the high rate of endemism
311 generally observed in subterranean obligate species (49) and parallels similar predictions obtained
312 for groundwater crustaceans in Europe (20). Although significant, the geographical effect was less
313 strong in the case of troglophile communities, consistently with their broad distribution patterns and
314 higher dispersal propensity. It seems likely that the steep increase in β -diversity with geographical
315 distance reflects dispersal limitations. In the case of communities at northern latitudes (> 48 – 50°
316 N), this pattern might also reflect Pleistocene local extirpation of faunas, and the subsequent post-
317 glacial dispersal limitation (50). However, at this analytical scale the influence of the distance from
318 LGM glaciers in our models was negligible—a variable often found to be highly significant to
319 explain the distribution of European subterranean arachnids at smaller scales (41,51).

320 The most important environmental gradient explaining dissimilarity in troglophile
321 communities was mean annual temperature. This variable was recovered as significant and

important also for troglobionts, although to a lesser degree. Mean annual temperature at the surface is deemed to be an ideal proxy-variable for the largely constant thermal conditions of subterranean habitats (40,52). The variable was also collinear with elevation, meaning that caves at different altitudes tend to be more dissimilar from one another. Insofar as the climatic distance between two caves explains the dissimilarity between their communities, it is possible to infer that the specialization to habitats with contrasted temperatures may have contributed to promote isolation (53). It is also worth noting that mean annual temperature strongly correlates with surface productivity; in turn, a high surface productivity is deemed to correlate with high organic input to subterranean habitats, thus potentially exerting an influence on diversity patterns (39,54,55). Interestingly, a few studies on groundwater fauna (20,47) recovered thermal seasonality as the most important factor explaining species range size, whereas temperature range was not significant in this study.

Finally, amount of karst was an additional important predictor for both troglophiles and troglobionts. Karst area is a good proxy for subterranean habitat availability and connectivity (34,39), so this result was somewhat expected. Yet, it is worth noting that 88% of records in the database were obtained from limestone caves. Thus, our analysis might underestimate the availability of suitable habitats to subterranean spiders in non-karst substrates, such as talus caves, shallow subterranean habitats (56), or artificial subterranean habitats opening in other substrates, all poorly represented in the dataset.

Interestingly, there was virtually no contribution of local cave features in explaining β -diversity patterns. We realize that this result might seem counterintuitive, because different geomorphological features have been documented to directly or indirectly correlate with subterranean diversity. For example, it is documented how a cave with a large entrance or a vertical cave with a high drop often accumulate more external trophic resources than a horizontal cave with a very narrow entrance (57), hence likely supporting a dissimilar and possibly more diverse community. Similarly, one might expect a cave with a greater planimetric development to support a more diverse community than a smaller cave. There are different explanations for this pattern. First, it is possible that local features exercise their primary effect on the species abundance, rather than on the simple presence/absence of species. Second, the effect of local variables may be evident exclusively locally; when analyzing diversity pattern at regional to continental scales, such effects may be masked by the stronger influence of large-scale gradients. Third, it is worth to note that caves are only part of an extended network of fissures which more properly represents the elective habitat of the subterranean fauna (58). The fact that we recovered a major effect of the karst area (that is, a proxy for the extent of habitat availability across the landscape), rather than that of the planimetric development of the cave (that is, an anthropocentric view of the habitat available to the subterranean species), suggests that this explanation might be reasonable. In a way, the lack of local effects is a possible clue reflecting our inability to truly

capture the local conditions governing subterranean habitats. Since the effect of local features turned out to be inconspicuous, caves emerge as ideal experimental model systems in which to study broad biological patterns without the confounding effect of local features.

362

363 CONCLUSIONS

364

Even though caves represent island-like habitats well-suited for α - and β -diversity studies (20,25,33,59–62), it is only recently that researchers began to consistently explore broad-scale patterns of subterranean diversity (21,63). We demonstrated a limited influence of local-scale cave features in determining the continental pattern of β -diversity in subterranean spider communities in Europe. On the other hand, we proved how geographical distance, in synergy with the environmental gradients of habitat and temperature, explained most of the community turnover. This pattern is consistent with the dispersal limitations that are typically observed in subterranean obligate species. Overall, this analysis was possible thanks to the collaboration of 31 researchers, providing their expertise and their own field-collected data. Such a collaborative attitude is a crucial premise to tackle macroecological issues, where the quality and the amount of data is an essential condition that is rarely met. This point is particularly important in cave-based science, as the harsh condition of the working environment delays the acquisition of the much-needed data for exploring global diversity patterns (21,63). Accordingly, we reaffirm the need to pursue collaborative databasing and data sharing (64).

379

380 Data accessibility

Response and environmental predictors, as well as the R code to generate the analyses, are available in Dryad (doi:10.5061/dryad.qz612jm8z). The full dataset of subterranean spider communities across Europe will be made available in an associated data paper (32), and updated as long as new data will become available.

385

386 Supporting information

Appendix S1. Geographical information collected for each locality, and local- and broad-scale predictors considered in the analyses.

389

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394

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406 **Author contribution**

SM and MI planned the line of enquiry. MI coordinated the network of experts. SM and PC analyzed the data. All other authors are listed alphabetically. AM and LK provided data from Slovakia. CD, MN, and SĆ provided data from Albania, Bulgaria, Greece, and Serbia. CK provided data from Austria. TB and SZ provided data from Germany. DA, LD, and GB provided data from Hungary. CEP, CR, and JF provided data from Spain. FG provided data from Slovenia. KBK and ME provided data from Turkey. MK and MP provided data from North Macedonia and Montenegro. MI and SM provided data from Italy. MP provided data from Croatia and Bosnia and Herzegovina. OTM provided data from Romania. PC provided data from Finland and Portugal. RR provided data from Poland. RSV provided data from Ukraine. VR provided data from the Czech Republic.

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585 **Tables**

586

587 **Table 1.** Relative importance and significance of noncollinear predictor variables for β -diversity of
588 subterranean spider communities across Europe, as determined by permutating 50 times the generalized
589 dissimilarity models (* $p < 0.05$). Em dashes indicate predictors which explained no model variance.

590

Variable	Scale	Troglophiles	Troglobionts
Geographical distance (°)	Broad	13.03 *	43.37 *
Cumulative precipitation (mm)	Broad	0.42	1.11
Mean annual temperature (°C)	Broad	48.96 *	11.38 *
Annual range of temperature (°C)	Broad	—	—
Distance from the LGM glacier (km)	Broad	—	—
Karst area (km ²)	Broad	16.74 *	5.36 *
N° of entrances	Local	—	—
Main entrance size (m)	Local	0.73	—
Cave development (m)	Local	—	0.44
Prevalent drop (m)	Local	0.02	2.25

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613 Figures captions

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615 **Figure 1.** Map of cave localities in Europe included in the dataset. Shades of grey represent elevation.
616 Brown areas indicate karst areas.

617

618 **Figure 2. (A)** Abundance classes (octaves) of the numbers of troglophile and troglobiont spider species in
619 caves included in the analysis. Error bars indicate 95% confidence limits for troglophile species ($n = 199$)
620 when re-sampled to the number of troglobiont species ($n = 132$). **(B)** Density of β -diversity values for
621 troglophile and troglobiont spider communities.

622

623 **Figure 3.** Results of generalized dissimilarity model for troglophile spider communities across Europe. **(A)**
624 Predicted spatial variation in troglophile spider community composition. Colors represent gradients in species
625 composition derived from transformed environmental predictors, whereby areas with similar colors are
626 expected to contain more similar communities. **(B)** Relation between observed compositional dissimilarity in
627 troglophile spider community between each cave pair and the linear predictor of the regression equation from
628 generalized dissimilarity model. **(C–E)** Fitted I-splines (partial regression fits) for variables significantly
629 associated with β -diversity of troglophile spiders. The maximum height reached by each curve indicates the
630 total amount of compositional turnover explained by that variable (holding all other variables constant),
631 whereas the shape of each spline indicates how the rate of compositional turnover varies along the
632 environmental gradient.

633

634 **Figure 4.** Results of generalized dissimilarity model for troglobiont spider communities across Europe. **(A)**
635 Predicted spatial variation in communities of troglobiont spiders. Colors represent gradients in species
636 composition derived from transformed environmental predictors, whereby areas with similar colors are
637 expected to contain more similar communities. Color gradient was constrained within a radius of 500 km
638 from all cave localities with troglobionts to avoid extending predictions to areas lacking data on troglobionts.
639 **(B)** Relation between observed compositional dissimilarity in troglobiont spider community between each
640 cave pair and the linear predictor of the regression equation from generalized dissimilarity model. **(C–E)**
641 Fitted I-splines (partial regression fits) for variables significantly associated with β -diversity of troglobiont
642 spiders. The maximum height reached by each curve indicates the total amount of compositional turnover
643 explained by that variable (holding all other variables constant), whereas the shape of each spline indicates
644 how the rate of compositional turnover varies along the environmental gradient.







